

Day I

RMT is a very vast subject, with many applications. To try to pretend to give lectures in 2 weeks would be selfish from my part. So, instead I am going to be selfish and present some of the work I have done during the last years that are could be entitled very broadly

Statistical Mechanics of Random Matrices

If you could go away with a main goal of these lectures will be: the art of ^{using} ~~stealing~~ tools of statistical mechanics (of disordered systems) to solve problems of interest in (in this case) Random matrices

Train of thought is the following

problem of interest in random matrices

map



problem in statistical mechanics of disordered systems (spin glasses)

solved by

using: * concepts of stat. mech

partition function
* free energies
* ...

* tools of disordered systems

* cavity method
* replica method

* some mathematical tools

* saddle point method
* multidimensional Gaussian integrals
* useful representations
Sol Dirac delta & Heaviside

Content of these lectures

- fundamentals
- 1st week
- I. Some useful mathematical tools | Day 1
 2. Tools from disordered systems | Days 2, 3
 3. Problems in RMs mapped into problems of statistical mechanics
 4. Particular problems → Days 4, 5

- 2nd week
- 4a) Spectral density of directed and undirected random graphs | days 1, 2
 - 4b) Large deviation theory for the spectrum of random graphs | days 3, 4

Days 5 → exam!

I Some mathematical tools

The saddle-point method

This method is also referred to it as steepest descent or Laplace method.

Suppose that:

$$\tilde{I}_N = \int_a^b dx e^{-Nf(x)}$$

f is a real function

study the asymptotic behaviour of \tilde{I}_N for large N. One can show that

$$\mathcal{I}_N \approx e^{-Nf(x_0)}$$

$$x_0: f'(x)|_{x=x_0} = 0$$

and this symbol means:

$$f''(x)|_{x=x_0} > 0$$

$$x_0 \in (a, b)$$

$$- \lim_{N \rightarrow \infty} \frac{1}{N} \ln \mathcal{I}_N = f(x_0)$$

Proof: Expand $f(x)$ around x_0

$$f(x) = f(x_0) + \frac{1}{2} f''(x_0) (x-x_0)^2 + \sum_{n \geq 3} \frac{1}{n!} f^{(n)}(x_0) (x-x_0)^n$$

this means that

$$\mathcal{I}_N = \int_a^b dx \exp[-Nf(x)]$$

$$= \int_a^b dx \exp \left[-Nf(x_0) - \frac{N}{2} |f''(x_0)| (x-x_0)^2 - \sum_{n \geq 3} \frac{N}{n!} f^{(n)}(x_0) (x-x_0)^n \right]$$

$$= e^{-Nf(x_0)} \int_a^b dx \exp\left(-\frac{N}{2} |f''(x_0)| (x-x_0)^2 + R(x)\right)$$

Do the following change of variables

$$y = \sqrt{N} (x-x_0)$$

$$= e^{-Nf(x_0)} \int_{\sqrt{N}(a-x_0)}^{\sqrt{N}(b-x_0)} \frac{dy}{\sqrt{N}} e^{-\frac{f''(x_0)}{2} y^2 + \tilde{R}(y)}$$

$$= e^{-Nf(x_0)} \int_{\sqrt{N}(a-x_0)}^{\sqrt{N}(b-x_0)} \frac{dy}{\sqrt{N}} e^{-\frac{f''(x_0)}{2} y^2} \sum_{n=0}^{\infty} \frac{\tilde{R}^n(y)}{n!}$$

Since $x_0 \in (a, b) \Rightarrow$

$$\left. \begin{array}{l} \sqrt{N}(b-x_0) \\ \sqrt{N}(a-x_0) \end{array} \right\} \rightarrow \left. \begin{array}{l} \infty \\ -\infty \end{array} \right\}$$

$$Z_N \approx e^{-Nf(x_0)} \int_{-\infty}^{\infty} \frac{dy}{\sqrt{N}} e^{-\frac{f''(x_0)}{2} y^2} \sum_{n=0}^{\infty} \frac{\tilde{R}^n(y)}{n!}$$

Gaussian measure + perturbative terms

Wick's theorem
(Feynman diagrams
etc.)

leading term

$$Z_N \approx e^{-Nf(x_0)} \sqrt{\frac{2\pi}{Nf''(x_0)}}$$

If f is complex function $f: \mathbb{C} \rightarrow \mathbb{C}$

$$Z_N = \int_{\gamma} dz e^{-Nf(z)}$$

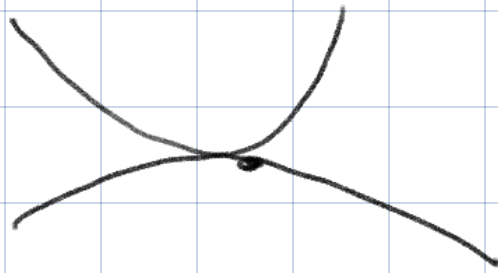
The same idea applies by deforming
the path γ so that it goes through
a point z_0 such that

$$f'(z_0) = 0$$

and then you further deform the path
so that the imaginary part of

$$-f''(z_0)(z-z_0)^2$$

is constant - for a while so that the Laplace method can be applied to the real part.



so that the Laplace method can be applied to the real part. The point at which $f'(z_0)=0$ has the form of a saddle (hence the name)

This is also called stationary phase approximation

The equation(s)/conditions for which

$$f'(x_0)=0$$

are called saddle-point equations

Of course this is quickly generalizable to multivariate case (and path integrals)

$$Z_N = \int d^N \vec{x} e^{-Nf(\vec{x})} \approx e^{-Nf(\vec{x}_0)}$$

with

$$\nabla f(\vec{x}) \Big|_{\vec{x}=\vec{x}_0} = 0$$

saddle
point
equations

and

$$\nabla_i \nabla_j f(\vec{x}) \Big|_{\vec{x}=\vec{x}_0} > 0$$

Hessian matrix

Multidimensional Gaussian integrals

Very important in condensed matter (FT, CFM, etc). In our case we worry about integral expressions of determinants

Suppose that we have a definite positive symmetric matrix A of size $N \times N$

$$A = A^T \quad T \text{ means to transpose}$$

and $A > 0$ (so eigenvalues are positive)

Then

$$\frac{1}{\sqrt{\det A}} = \int_{\mathbb{R}^N} \frac{d^N \vec{x}}{(2\pi)^{N/2}} e^{-\frac{1}{2} \vec{x}^T A \vec{x}}$$
$$d^N \vec{x} = \prod_{i=1}^N dx_i$$

or more generally

$$\frac{1}{\sqrt{|\det A|}} e^{\frac{i}{2} \vec{b}^T A \vec{b}} = \int_{\mathbb{R}^N} \frac{d^N \vec{x}}{(2\pi)^{N/2}} e^{\frac{i}{2} \vec{x}^T A \vec{x} + \vec{b}^T \vec{x}}$$

~~But~~, this is also called Hubbard-Stratonovich transformation

In physics these are usually called generating or external fields

Proof start with $\vec{b} = \vec{0}$. Since $A = A^T$
then

$$\Delta = O^T A O \quad O \equiv \text{orthogonal transformation}$$

$$\text{and } \Delta = \text{diag}(d_1, \dots, d_N)$$

$$\text{since } A > 0 \Rightarrow d_i > 0 \quad \forall i = 1, \dots, N$$

Define a transformation $\vec{x} \rightarrow \vec{x}'$

$$\vec{x} = O \vec{x}'$$

this implies that

$$\begin{aligned}\vec{x}^T A \vec{x} &= \vec{x}^T O^T A O \vec{x}' = \\ &= \vec{x}'^T \Delta \vec{x}' = \sum_{i=1}^N d_i x_i'^2\end{aligned}$$

also $d^N \vec{x} = \left| \frac{\partial \vec{x}}{\partial \vec{x}'} \right| d^N \vec{x}'$

↳ Jacobian of transformation

But $\frac{\partial \vec{x}}{\partial \vec{x}'} = O$ and $\det O = \pm 1$

Therefore

$$\begin{aligned}&\int_{\mathbb{R}^N} d^N \vec{x} e^{-\frac{1}{2} \vec{x}^T A \vec{x}} \\ &= \int_{\mathbb{R}^N} d^N \vec{x}' e^{-\frac{1}{2} \sum_{i=1}^N d_i x_i'^2} \\ &= \prod_{i=1}^N \int dx e^{-d_i x^2 / 2} = \prod_{i=1}^N \sqrt{\frac{2\pi}{d_i}}\end{aligned}$$

$$i=1 \quad \text{OK} \quad \quad \quad i=1$$

$$= \frac{(2\pi)^{N/2}}{\sqrt{\prod_{i=1}^N d_i}} = \frac{(2\pi)^{N/2}}{\det \Delta}$$

$$= \frac{(2\pi)^{N/2}}{\det A}, \quad \text{since } \det \Delta = \det A$$

Exercise. Do the case for $\vec{b} \neq \vec{0}$

Suppose now that A is an $N \times N$ complex matrix with nonvanishing determinant.

Then

$$\frac{1}{\det A} = \int \prod_{i=1}^N \frac{dz_i d\bar{z}_i}{2\pi i} \exp\left[-\sum_{i,j=1}^N \bar{z}_i A_{ij} z_j\right]$$

with $z_i, \bar{z}_i \in \mathbb{C}$

or more generally

$$\frac{\exp\left(\sum_{i,j=1}^N \bar{b}_i A_{ij} b_j\right)}{\det A} =$$

$$= \int \left[\prod_{i=1}^N \frac{dz_i d\bar{z}_i}{2\pi i} \right] \exp \left[- \sum_{i,j=1}^N \bar{z}_i A_{ij} z_j + \sum_i (b_i \bar{z}_i + \bar{b}_i z_i) \right]$$

these are generating fields

Exercise: prove this (Book of Zinn-Justin)

Other useful results for our mappings

Schottky-Plemelj identity

$$\lim_{z \rightarrow 0^+} \frac{1}{x - iz} = \mathcal{P} \frac{1}{x} + i\pi \delta(x)$$

and a prescription for the Heaviside step function

$$\Theta(-x) = \lim_{z \rightarrow 0^+} \frac{1}{2\pi i} \left[\log(x+iz) - \log(x-iz) \right]$$